Partical Swarm Optimization

Application: Traveling sales Man import numpy as np #Function to calculate the total distance of a route (path) def calculate_total_distance(route, distance_matrix): $total_distance = 0$ for i in range(len(route) - 1): total_distance += distance_matrix[route[i], route[i + 1]] total_distance += distance_matrix[route[-1], route[0]] # Return to start return total_distance # Particle Swarm Optimization (PSO) for TSP class PSO_TSP: def __init__(self, distance_matrix, num_particles=30, num_iterations=100, w=0.5, c1=1, c2=1): self.num_particles = num_particles self.num_iterations = num_iterations self.distance_matrix = distance_matrix self.num_cities = len(distance_matrix) self.w = w # Inertia weight self.c1 = c1 # Cognitive coefficientself.c2 = c2 # Social coefficient# Initialize particles' positions (routes) and velocities self.particles = np.array([np.random.permutation(self.num cities) for in range(num particles)]) self.velocities = np.array([np.zeros(self.num cities) for in range(num particles)]) #Evaluate fitness of each particle (route) self.fitness = np.array([calculate total distance(route, distance matrix) for route in self.particles]) # Initialize personal best positions and fitness self.p best = np.copy(self.particles) self.p best fitness = np.copy(self.fitness)

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# Initialize global best position and fitness
     self.g_best = self.p_best[np.argmin(self.p_best_fitness)]
     self.g_best_fitness = np.min(self.p_best_fitness)
  # Update velocities and positions
  def update_particles(self):
     for i in range(self.num_particles):
         # Update velocity: w * velocity + c1 * random() * (personal best - current position) + c2
* random() * (global best - current position)
       r1 = np.random.rand(self.num cities)
       r2 = np.random.rand(self.num cities)
       cognitive velocity = self.c1 * r1 * (self.p best[i] - self.particles[i])
       social_velocity = self.c2 * r2 * (self.g_best - self.particles[i])
       inertia velocity = self.w * self.velocities[i]
       self.velocities[i] = inertia_velocity + cognitive_velocity + social_velocity
       # To ensure we move to a new route, modify the velocity to shuffle positions
       velocity_order = np.argsort(self.velocities[i]) # Sort based on the velocity magnitude
       new_particle = np.array([self.particles[i][j] for j in velocity_order])
       #Ensure the new particle is a valid permutation
       self.particles[i] = new particle
       self.fitness[i] = calculate_total_distance(new_particle, self.distance_matrix)
       # Update personal best
       if self.fitness[i] < self.p_best_fitness[i]:
          self.p_best[i] = self.particles[i]
          self.p_best_fitness[i] = self.fitness[i]
       # Update global best
       if self.fitness[i] < self.g_best_fitness:
          self.g_best = self.particles[i]
          self.g best fitness = self.fitness[i]
  # Run the PSO algorithm
  def run(self):
     for iteration in range(self.num_iterations):
       self.update_particles()
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print(f"Iteration {iteration + 1}: Best Distance = {self.g best fitness}")
     return self.g best, self.g best fitness
# Function to take user input for distance matrix and PSO parameters
def input_pso_parameters():
  # Input the number of cities and distance matrix
  num_cities = int(input("Enter the number of cities: "))
  print("Enter the distance matrix row by row (space-separated):")
  distance matrix = np.zeros((num cities, num cities))
  for i in range(num cities):
     row = list(map(int, input(f"Row {i + 1}: ").split()))
    distance matrix[i] = row
  # Input PSO parameters
  num_particles = int(input("Enter the number of particles: "))
  num_iterations = int(input("Enter the number of iterations: "))
  w = float(input("Enter the inertia weight (w): "))
  c1 = float(input("Enter the cognitive coefficient (c1): "))
  c2 = float(input("Enter the social coefficient (c2): "))
  return distance_matrix, num_particles, num_iterations, w, c1, c2
#Get user input for the distance matrix and PSO parameters
distance_matrix, num_particles, num_iterations, w, c1, c2 = input_pso_parameters()
# Initialize PSO with the distance matrix and parameters
pso tsp = PSO TSP(distance matrix, num particles, num iterations, w, c1, c2)
#Run PSO to find the shortest path
best_route, best_distance = pso_tsp.run()
print("\nBest route found:", best_route)
print("Best route distance:", best_distance)
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Output:

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Enter the number of cities: 4
Enter the distance matrix row by row (space-separated
Row 1: 0 5 10 15
Row 2: 5 0 20 30
Row 3: 30 10 0 5
Row 4: 5 10 15 0
Enter the number of particles: 50
Enter the number of iterations: 200
Enter the inertia weight (w): 0.7
Enter the cognitive coefficient (c1): 1.5
Enter the social coefficient (c2): 1.5
Iteration 1: Best Distance = 30.0
Iteration 2: Best Distance = 30.0
Iteration 3: Best Distance = 30.0
Iteration 4: Best Distance = 30.0
Iteration 5: Best Distance = 30.0
Iteration 6: Best Distance = 30.0
Iteration 7: Best Distance = 30.0
Iteration 8: Best Distance = 30.0
Iteration 185: Best Distance = 30.0
Iteration 186: Best Distance = 30.0
Iteration 187: Best Distance = 30.0
Iteration 188: Best Distance = 30.0
Iteration 189: Best Distance = 30.0
Iteration 190: Best Distance = 30.0
Iteration 191: Best Distance = 30.0
Iteration 192: Best Distance = 30.0
Iteration 193: Best Distance = 30.0
Iteration 194: Best Distance = 30.0
Iteration 195: Best Distance = 30.0
Iteration 196: Best Distance = 30.0
Iteration 197: Best Distance = 30.0
Iteration 198: Best Distance = 30.0
Iteration 199: Best Distance = 30.0
Iteration 200: Best Distance = 30.0
Best route found: [2 3 1 0]
Best route distance: 30.0
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